

Planning

A planning agent will construct plans to achieve its goals, and then execute them.

Analyze a situation in which it finds itself and develop a strategy for achieving the agent's goal.

Achieving a goal requires finding a sequence of actions that can be expected to have the desired outcome.

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Problem Solving

Representation of actions – actions generate successor states

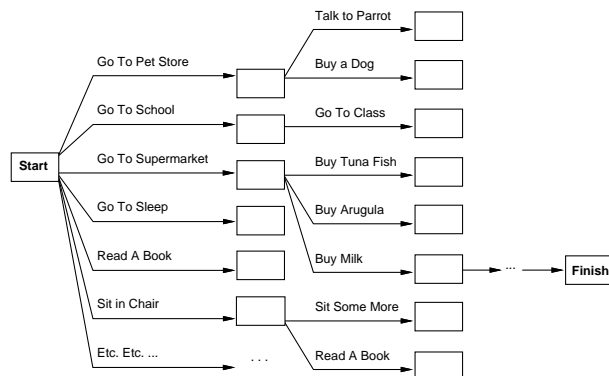
Representation of states – all state representations are complete

Representation of goals – contained in goal test and heuristic function

Representation of plans – unbroken sequence of actions leading from initial to goal state.

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Planning Example



GOAL: Get a quart of milk and a bunch of bananas and a variable-speed cord-less drill.

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Planning Versus Problem Solving

(1) Open up the representation of states, goals and actions.

- States and goals represented by sets of sentences – *Have(Milk)*
- Actions represented by rules that represent their preconditions and effects: *Buy(x)* achieves *Have(x)*

This allows the planner to make direct connections between states and actions.

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Planning Versus Problem Solving

(2) Planner is free to add actions to the plan wherever they are needed, rather than in an incremental sequence starting at the initial state.

- No connection between the order of planning and the order of execution.
- Representation of states as sets of logical sentences makes this freedom possible.

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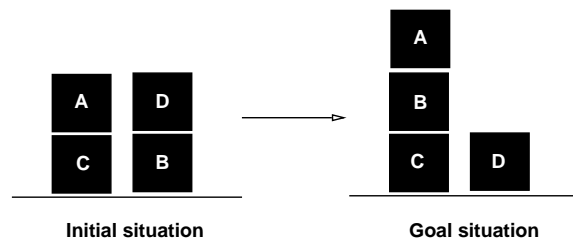
Planning Versus Problem Solving

(3) Most parts of the world are independent of most other parts.

- Can solve $Have(Milk) \wedge Have(Bananas) \wedge Have(Drill)$ using divide-and-conquer strategy.
- Can re-use subplans (go to supermarket)

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Planning as a Logical Inference Problem



Axioms:

$On(A,C), On(C,Table), On(D,B), On(B,Table), Clear(A),$
 $Clear(D)$

plus rules for moving things around...

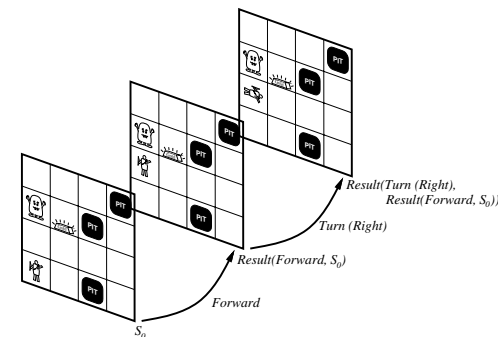
Prove: $On(A,B) \wedge On(B,C)$

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Planning as Deduction: Situation Calculus

In first-order logic, once a statement is shown to be true, it remains true forever.

Situation calculus: way to describe change in first-order logic.



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Situation Calculus

fluents: functions and predicates that vary from one situation to the next

$on(A, C)$ $on(A, C, S_0)$

$at(agent, [1, 1])$ $at(agent, [1, 1], S_0)$

atemporal functions and predicates are also allowed

$block(A)$

$gold(G_1)$

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Situation Calculus: Actions

Actions are described by stating their effects.

possibility axiom: $preconditions \Rightarrow Poss(a, s)$.

$\forall s \forall x \neg On(x, Table, s) \wedge Clear(x, s) \Rightarrow$
 $Poss(PlaceOnTable(x), s)$

effect axiom: $Poss(a, s) \Rightarrow$ Changes that result from taking the action.

$\forall s \forall x Poss(PlaceOnTable(x), s) \Rightarrow$
 $On(x, Table, Result(PlaceOnTable(x), s))$

$\forall s \forall y \forall z On(y, z, s) \wedge (z \neq Table) \Rightarrow \neg On(y, Table, s)$

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Situation Calculus: Action Sequences

$Result([], s) = s$.

$Result([a|seq], s) = Result(seq, Result(a, s))$.

We'd like to be able to prove:

$\exists seq \ On(A, B, Result(seq, S_0)) \wedge On(B, C, Result(seq, S_0))$

which would produce, for example, the following:

$On(A, B, Result([PoT(A), PoT(D), Put(B, C), Put(A, B)], S_0))$
 $\wedge On(B, C, Result([PoT(A), PoT(D), Put(B, C), Put(A, B)], S_0))$

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The Frame Problem

Actions don't specify what happens to objects not involved in the action, but the logic framework requires that information.

$\forall s \forall x Poss(PoT(x), s) \Rightarrow On(x, Table, Result(PoT(x), s))$

Frame axioms: Inform the system about preserved relations.

$\forall s \forall x \forall y \forall z [(On(x, y, s) \wedge (x \neq z)) \Rightarrow On(x, y, Result(PoT(z), s))]$

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... and Its Relatives

representational frame problem: proliferation of frame axioms.

Solution: use **successor-state axioms**

Action is possible \Rightarrow (Fluent is true in result state \Leftrightarrow (Action's effect made it true \vee It was true before and action left it alone)).

inferential frame problem: have to carry each property through all intervening situations during problem-solving, even if the property remains unchanged throughout

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qualification problem: difficult, in the real world, to define the circumstances under which a given action is guaranteed to work

ramification problem: proliferation of *implicit* consequences of actions.

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The Need for Special Purpose Algorithms

So...We have a formalism for expressing goals and plans and we can use resolution theorem proving to find plans.

Problems:

- frame problem
- time to find plan can be exponential
- logical inference is semi-decidable
- resulting plan could have many irrelevant steps

We'll need to:

- restrict language
- use a special purpose algorithm called a planner

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The STRIPS Language

States and Goals: Conjunctions of positive, function-free literals. No variables.

Have (Milk) \wedge Have (Bananas) \wedge Have (Drill)
 \wedge At (Home)

Closed world assumption: any conditions that are not mentioned in a state are assumed false.

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Actions:

preconditions: conjunction of positive, function-free literals that must be true before the operator can be applied.

effects: conjunction of function-free literals; *add* list and *delete* list.

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STRIPS assumption

Every literal not mentioned in the effect remains unchanged in the resulting state when the action is executed.

Avoids the representational frame problem.

Solution for the planning problem: an action sequence that, when executed in the initial state, results in a state that satisfies the goal.

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STRIPS Actions

Move block x from block y to block z

preconds: $On(x, y) \wedge Block(x) \wedge Block(z)$
 $\wedge Clear(x) \wedge Clear(z)$

effects: Add: $On(x, z)$, $Clear(y)$
Delete: $On(x, y)$, $Clear(z)$

Move block x from block y to Table

preconds: $On(x, y) \wedge Block(x) \wedge Block(y) \wedge Clear(x)$
effects: Add: $On(x, Table)$, $Clear(y)$
Delete: $On(x, y)$

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Move block x from Table to block z

preconds: $On(x, Table) \wedge Block(x) \wedge Block(z)$
 $\wedge Clear(x) \wedge Clear(z)$

effects: Add: $On(x, z)$
Delete: $On(x, Table)$, $Clear(z)$

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Plan by Searching for a Satisfactory Sequence of Actions

progression planner searches forward from the initial situation to the goal situation

regression planner search backwards from the goal state to the initial state

Heuristics: derive a **relaxed problem**; employ the **subgoal independence** assumption.

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Searching Plan Space

Alternative is to search through the space of *plans* rather than the original state space.

Start with simple, incomplete **partial plan**; expand until complete.

Operators: add a step, impose an ordering on existing steps, instantiate a previously unbound variable.

Refinement Operators take a partial plan and add constraints

Modification Operators are anything that is not a refinement operator; take an incorrect plan and debug it.

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Representation for Plans

Goal: $RightShoeOn \wedge LeftShoeOn$

Initial state: λ

Operators:

Action	Preconds	Effect
RightShoe	RightSockOn	RightShoeOn
RightSock	λ	RightSockOn
LeftShoe	LeftSockOn	LeftShoeOn
LeftSock	λ	LeftSockOn

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Partial Plans

Partial Plan: RightShoe LeftShoe

Principle of **Least Commitment** says to only make choices about things that you currently care about.

Partial order planner – can represent plans in which some steps are ordered and others are not.

Total order planner considers a plan a simple list of steps

A **linearization of P** is a totally ordered plan that is derived from a plan P by adding ordering constraints.

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Defer Variable Binding

Planners must commit to bindings for variables

Example: Goal: Have(Milk) Action: Buy(item,store)

Principle of Least Commitment: Only make choices about things that you care about, leaving other details to be worked out later.

Buy(Milk,K-MART) versus Buy(Milk,store)

Fully instantiated plan: every variable is bound to a constant.

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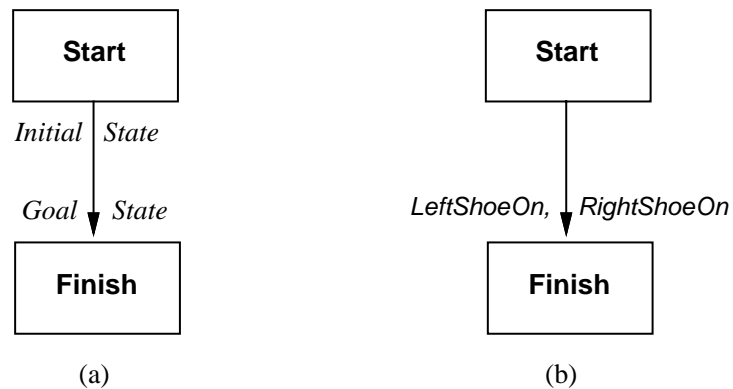
Definition of a Plan

- A set of plan steps (actions).
- A set of step ordering constraints of the form $S_i \prec S_j$
- A set of variable binding constraints
- A set of causal links, written as $S_i \xrightarrow{c} S_j$

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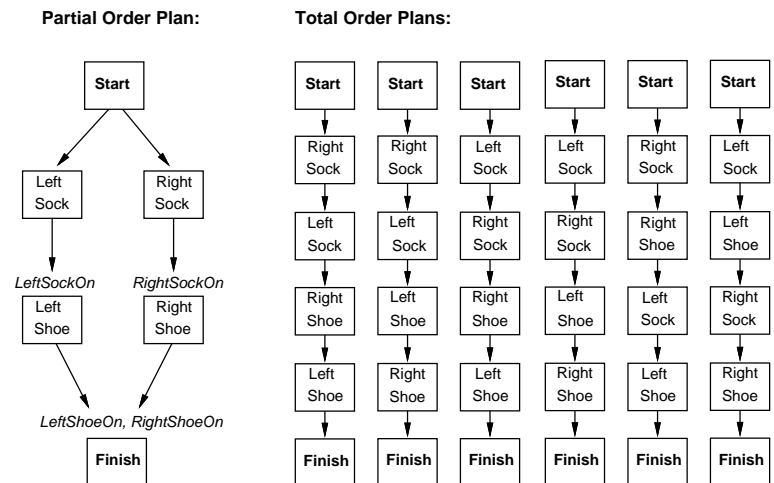
Initial Plan for Shoes and Socks

Initial plan: $Start \prec Finish$



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Partial Plan for Shoes and Socks



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Planner Output

A solution is a complete, consistent plan.

A complete plan: every precondition of every step is achieved by some other step.

A consistent plan: there are no contradictions in the ordering or binding constraints. Contradiction occurs when both $S_i \prec S_j$ and $S_j \prec S_i$.

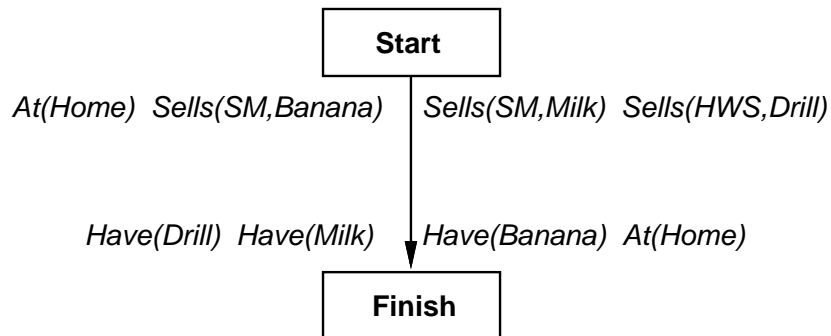
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POP Example: STRIPS Actions

Action	PreCond	Effect
$Go(there)$	$At(there)$	$At(there) \wedge \neg At(here)$
$Buy(x)$	$At(store) \wedge Sells(store, x)$	$Have(x)$

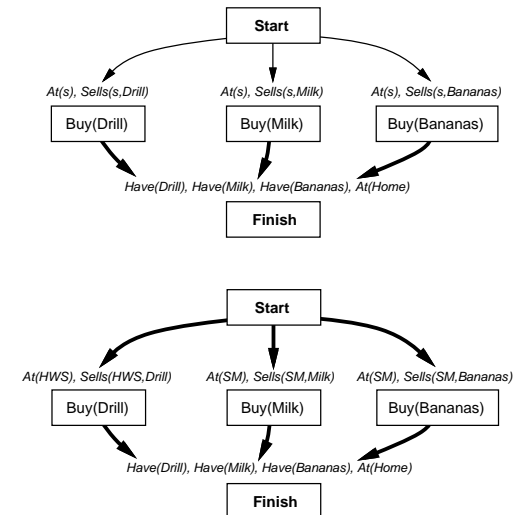
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POP Example: Initial Plan

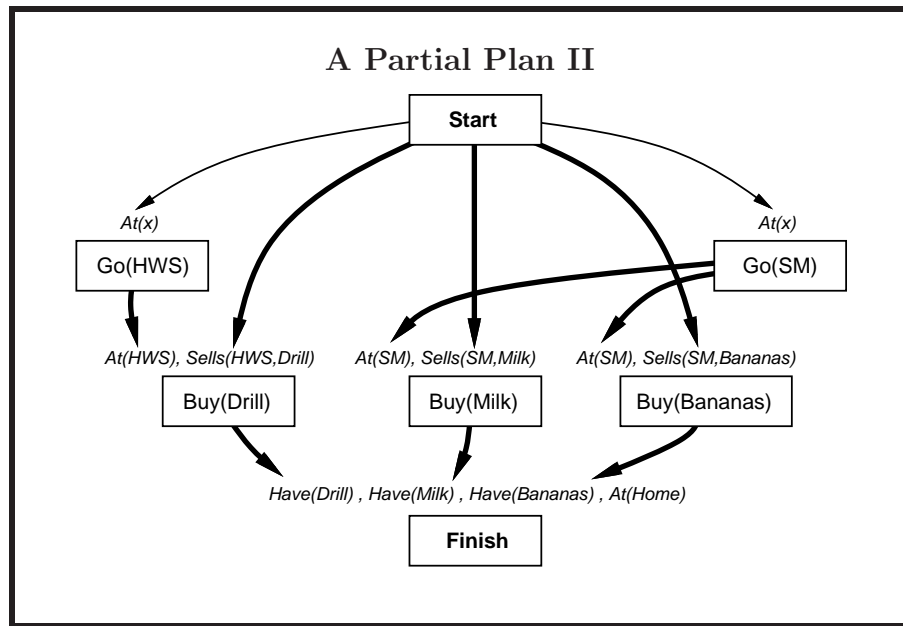


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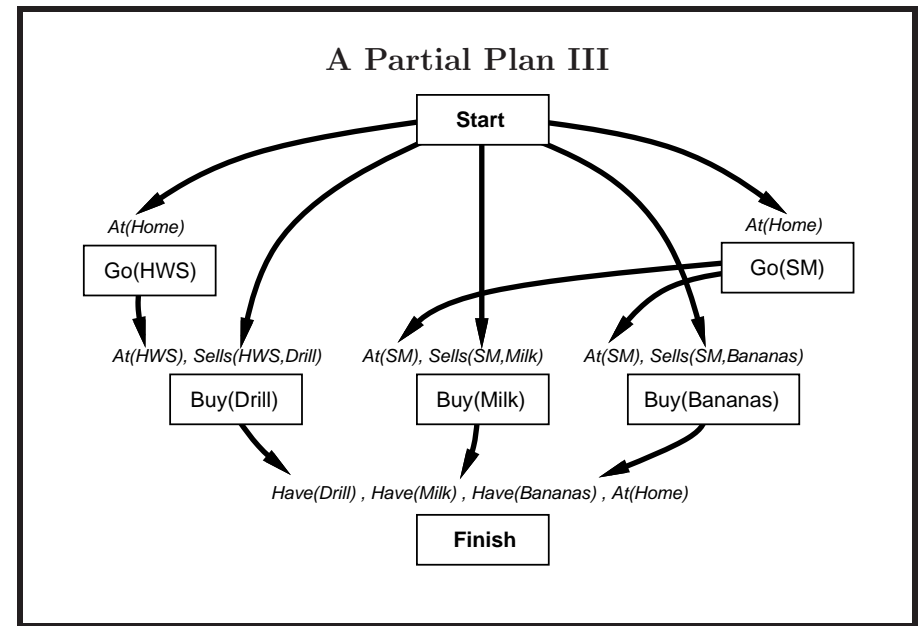
A Partial Plan I



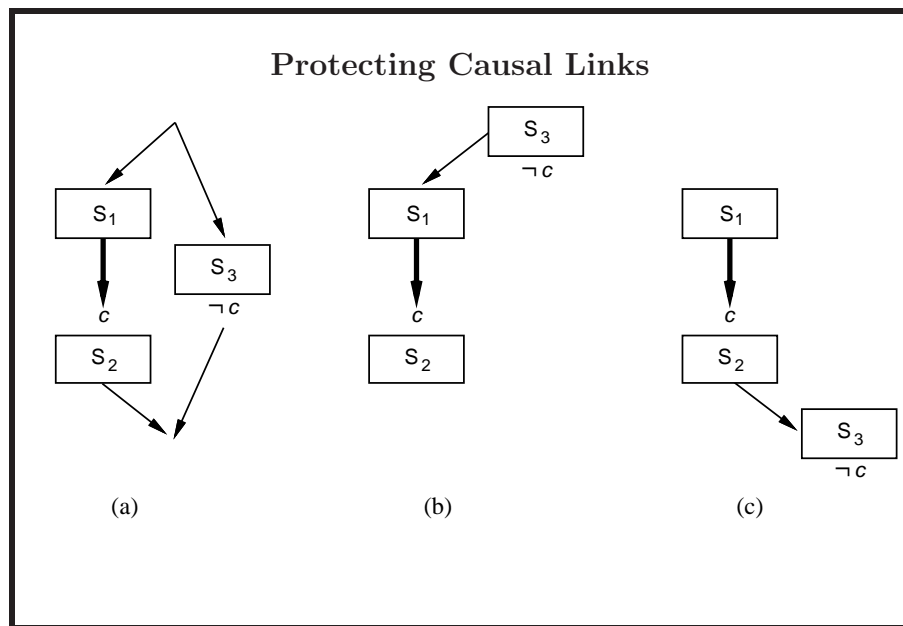
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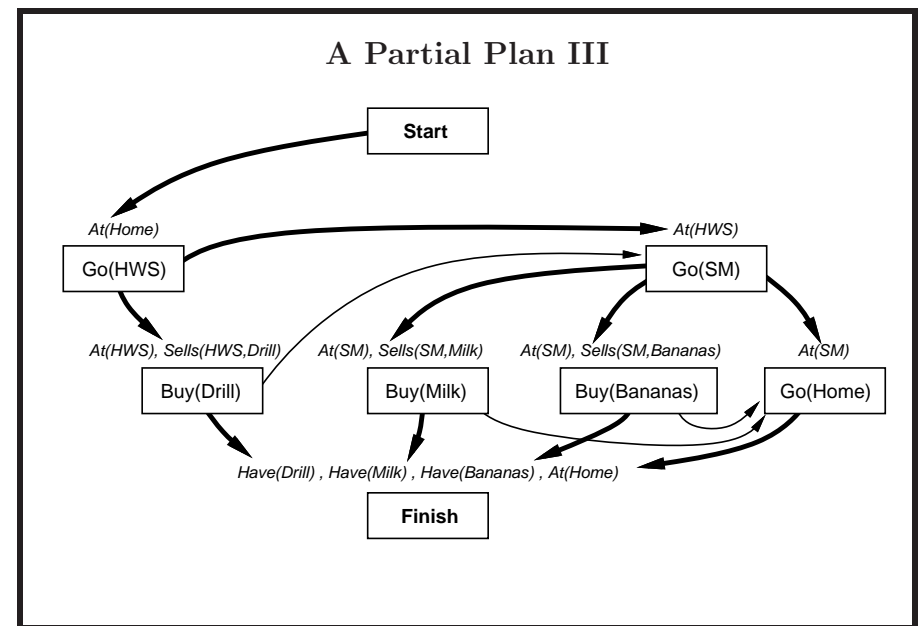
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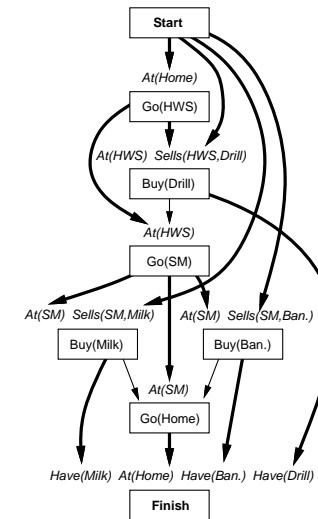
Achieving At(Home)

Candidate link	Threats
At(x) to initial state	Go(HWS), Go(SM)
At(x) to Go(HWS)	Go(SM)
At(x) to Go(SM)	At(SM) preconds of Buy(Milk), Buy(Bananas)

Solution: Link At(x) to Go(SM), but order Go(Home) to come after Buy(Bananas) and Buy(Milk).

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A final Plan



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```

function POP(initial, goal, operators) returns plan
    plan ← MAKE-MINIMAL-PLAN(initial, goal)
    loop do
        if SOLUTION?(plan) then return plan
        Sneed, c ← SELECT-SUBGOAL(plan)
        CHOOSE-OPERATOR(plan, operators, Sneed, c)
        RESOLVE-THREATS(plan)
    end

function SELECT-SUBGOAL(plan) returns Sneed, c
    pick a plan step Sneed from STEPS(plan)
    with a precondition c that has not been achieved
    return Sneed, c

procedure CHOOSE-OPERATOR(plan, operators, Sneed, c)
    choose a step Sadd from operators or STEPS(plan) that has c as an effect
    if there is no such step then fail
    add the causal link Sadd → Sneed to LINKS(plan)
    add the ordering constraint Sadd < Sneed to ORDERINGS(plan)
    if Sadd is a newly added step from operators then
        add Sadd to STEPS(plan)
        add Start < Sadd < Finish to ORDERINGS(plan)

procedure RESOLVE-THREATS(plan)
    for each Sthreat that threatens a link Si → Sj in LINKS(plan) do
        choose either
            Promotion: Add Sthreat < Si to ORDERINGS(plan)
            Demotion: Add Sj < Sthreat to ORDERINGS(plan)
        if not CONSISTENT(plan) then fail
    end
  
```

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Strengths of Partial-Order Planning Algorithms

- Takes a huge state space problem and solves in only a few steps.
- Least commitment strategy means that search only occurs in places where sub-plans interact.
- Causal links allow planner to recognize when to abandon a doomed plan without wasting time exploring irrelevant parts of the plan.

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Practical Planners

STRIPS approach is insufficient for many practical planning problems. Can't express:

resources: Operators should incorporate resource consumption and generation. Planners have to handle constraints on resources efficiently.

time: Real-world planners need a better model of time.

hierarchical plans: Need the ability to specify plans at varying levels of detail.

Also need to incorporate heuristics for guiding search.

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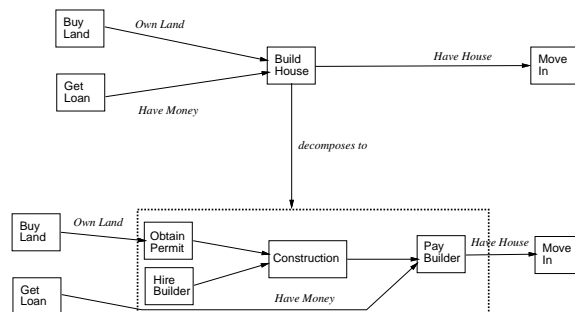
Planning Graphs

- Data structure (graphs) that represents plans, and can be efficiently constructed, and that allows for better heuristic estimates.
- **Graphplan:** algorithm that processes the planning graph, using backward search, to extract a plan.
- **SATPlan:** algorithm that translates a planning problem into propositional axioms and applies a CSP algorithm to find a valid plan.

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Hierarchical Planning



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Spacecraft Assembly, integration and verification (AIV)

- OPTIMUM-AIV used by the European Space Agency to AIV spacecraft.
- Generates plans and monitors their execution – ability to re-plan is the principle objective.
- Uses O-Plan architecture – like partial-order planner, but can represent time, resources and hierarchical plans. Accepts heuristics for guiding search and records its reasons for each choice.

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Scheduling for Space Missions

- Planners have been used by the ground teams for the Hubble space telescope and for the Voyager, Uosat-II and ERS-1.
- Goal: coordinate the observational equipment, signal transmitters and altitude and velocity-control mechanism in order to maximize the value of the information gained from observations while obeying resource constraints on time and energy.

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